

# 3.1

# Transcriptomic-isoforms

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# Working directory

How are our folders structured

- ▼ 3.1\_Transcriptomics
  - ▶ basecall
  - ▶ mapping
  - ▶ outputs
  - ▶ SUPPA2
- ▼ 3.2\_Epitranscriptomics
  - ▶ basecall
  - ▶ outputs
- ▼ Arabidopsis\_references
  - ▶ genome
  - ▶ transcriptome
- ▼ ont\_raw\_data
  - 📄 sample\_20reads.pod5
- ▼ scripts
  - 📄 salmon\_script\_ONT.sh
- ▼ tools
  - ▶ minimap2-2.30\_x64-linux
  - ▶ SUPPA-2.4

# A little bit about the samples

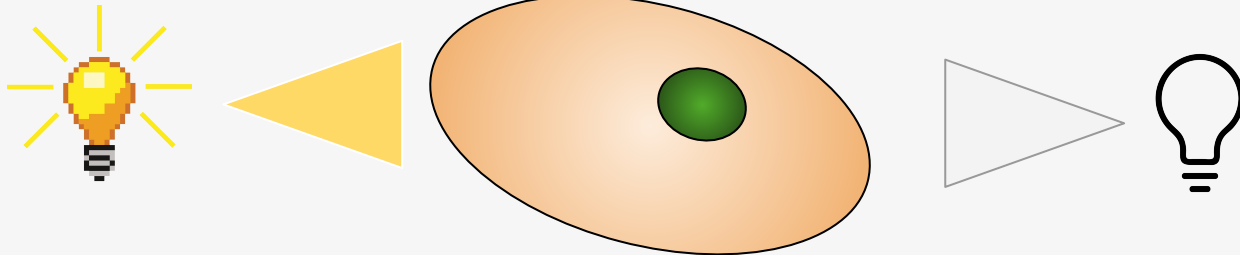
(BARCODES)

**1 – 3** **Nuclear Fraction**  
Light treatment

**4 – 6** **Nuclear Fraction**  
Dark treatment

**7 – 9** **Citoplasm**  
Light treatment

**10 – 12** **Citoplasm**  
Dark treatment



# Dorado Basecaller



```
(rna_workshop) lucas-elytron@Lucas-Elytron:~$ dorado -h
[2025-07-14 17:04:30.588] [info] Running: "-h"
Usage: dorado [options] subcommand

Positional arguments:
aligner
basecaller
correct
demux
download
duplex
polish
summary
trim

Optional arguments:
-h --help          shows help message and exits
-v --version       prints version information and exits
```

# Dorado Basecaller



```
(rna_workshop) lucas-elytron@lucas-elytron: $ dorado basecaller -h
[2025-07-14 17:06:39.960] [info] Running: "basecaller" "-h"
Usage: dorado [--help] [--verbose]... [--device VAR] [--models-directory VAR] [--bed-file VAR] [--recursive] [--read-ids VAR] [--max-reads VAR] [--resume-from VAR] [--min-qscore VAR] [--emit-moves] [--emit-fastq] [--emit-sam] [--output-dir VAR] [--reference VAR] [--mm2-opts VAR] [--modified-bases VAR...] [--modified-bases-models VAR] [--modified-bases-threshold VAR] [--modified-bases-batchsize VAR] [--kit-name VAR] [--sample-sheet VAR] [--barcode-both-ends] [--barcode-arrangement VAR] [--barcode-sequences VAR] [--primer-sequences VAR] [--no-trim] [--trim VAR] [--estimate-poly-a] [--poly-a-config VAR] [--batchsize VAR] [--chunksize VAR] [--overlap VAR] model data

Positional arguments:
  model          Model selection {fast,hac,sup}@v{version} for automatic model selection including modbases, or path to existing model directory.
  data           The data directory or file (PODS/FAST5 format).

Optional arguments:
  -h, --help          shows help message and exits
  -v, --verbose        [may be repeated]
  -x, --device        Specify CPU or GPU device: 'auto', 'cpu', 'cuda:all' or 'cuda:<device_id>[,<device_id>...]. Specifying 'auto' will choose either 'cpu', 'metal' or 'cuda:all' depending on the presence of a GPU device. [nargs=0..1] [default: "auto"]
  --models-directory  Optional directory to search for existing models or download new models into. [nargs=0..1] [default: "."]
  --bed-file          Optional bed-file. If specified, overlaps between the alignments and bed-file entries will be counted, and recorded in BAM output using the 'bh' read tag. [nargs=0..1] [default: ""]

Input data arguments (detailed usage):
  -r, --recursive      Recursively scan through directories to load FAST5 and POD5 files.
  -l, --read-ids       A file with a newline-delimited list of reads to basecall. If not provided, all reads will be basecalled. [nargs=0..1] [default: ""]
  -n, --max-reads      Limit the number of reads to be basecalled. [nargs=0..1] [default: 0]
  --resume-from        Resume basecalling from the given HTS file. Fully written read records are not processed again. [nargs=0..1] [default: ""]

Output arguments (detailed usage):
  --min-qscore         Discard reads with mean Q-score below this threshold. [nargs=0..1] [default: 0]
  --emit-moves         Write the move table to the 'mv' tag.
  --emit-fastq         Output in fastq format.
  --emit-sam           Output in SAM format.
  -o, --output-dir     Optional output folder, if specified output will be written to a calls file (calls_<timestamp>.sam|.bam|.fastq) in the given folder.

Alignment arguments (detailed usage):
  --reference           Path to reference for alignment. [nargs=0..1] [default: ""]
  --mm2-opts           Optional minimap2 options string. For multiple arguments surround with double quotes.

Modified model arguments (detailed usage):
  --modified-bases     A space separated list of modified base codes. Choose from: pseU, m6A_DRACH, m6A, 6mA, m5C, 5mC, 5mCG_5hmCG, 5mCG, 5mC_5hmC, inosine_m6A, 4mC_5mC. [nargs: 1 or more]
  --modified-bases-models A comma separated list of modified base model paths. [nargs=0..1] [default: ""]
  --modified-bases-threshold The minimum predicted methylation probability for a modified base to be emitted in an all-context model, [0, 1].
  --modified-bases-batchsize The modified base models batch size.

Barcoding arguments (detailed usage):
  --kit-name           Enable barcoding with the provided kit name. Choose from: EXP-NBD103 EXP-NBD104 EXP-NBD114 EXP-NBD114-24 EXP-NBD196 EXP-PBC001 EXP-PBC096 SQK-16S024 SQK-16S114-24 SQK-LWB001 SQK-MAB114-24 SQK-MLK111-96-XL SQK-MLK114-96-XL SQK-NBD111-24 SQK-NBD111-96 SQK-NBD114-24 SQK-NBD114-96 SQK-PBK004 SQK-PCB109 SQK-PCB110 SQK-PCB111-24 SQK-PCB114-24 SQK-RAB201 SQK-RAB204 SQK-RBK001 SQK-RBK004 SQK-RBK110-96 SQK-RBK111-24 SQK-RBK111-96 SQK-RBK114-24 SQK-RBK114-96 SQK-RLB001 SQK-RPB004 SQK-RPB114-24 TWIST-16-UDI TWIST-96A-UDI VSK-PTC001 VSK-VMK001 VSK-VMK004 VSK-VPS001. [nargs=0..1] [default: ""]
  --sample-sheet       Path to the sample sheet to use. [nargs=0..1] [default: ""]
```

# Mapping the reads – Minimap2

## Getting Started

```
git clone https://github.com/lh3/minimap2
cd minimap2 && make
# long sequences against a reference genome
./minimap2 -a test/MT-human.fa test/MT-orang.fa > test.sam
# create an index first and then map
./minimap2 -x map-ont -d MT-human-ont.mmi test/MT-human.fa
./minimap2 -a MT-human-ont.mmi test/MT-orang.fa > test.sam
# use presets (no test data)
./minimap2 -ax map-pb ref.fa pacbio.fq.gz > aln.sam           # PacBio CLR genomic reads
./minimap2 -ax map-ont ref.fa ont.fq.gz > aln.sam            # Oxford Nanopore genomic reads
./minimap2 -ax map-hifi ref.fa pacbio-ccs.fq.gz > aln.sam    # PacBio HiFi/CCS genomic reads (v2.1+)
./minimap2 -ax lr:hq ref.fa ont-Q20.fq.gz > aln.sam          # Nanopore Q20 genomic reads (v2.27+)
./minimap2 -ax sr ref.fa read1.fa read2.fa > aln.sam          # short genomic paired-end reads
./minimap2 -ax splice ref.fa rna-reads.fa > aln.sam           # spliced long reads (strand unknown)
./minimap2 -ax splice -uf -k14 ref.fa reads.fa > aln.sam      # noisy Nanopore direct RNA-seq
./minimap2 -ax splice:hq -uf ref.fa query.fa > aln.sam        # PacBio Kinnex/Iso-seq (RNA-seq)
./minimap2 -ax splice --junc-bed=anno.bed12 ref.fa query.fa > aln.sam # use annotated junction
./minimap2 -ax splice:sr ref.fa r1.fq r2.fq > aln.sam          # short-read RNA-seq (v2.29+)
./minimap2 -ax splice:sr -j anno.bed12 ref.fa r1.fq r2.fq > aln.sam
./minimap2 -cx asm5 asm1.fa asm2.fa > aln.paf                # intra-species asm-to-asm alignment
./minimap2 -x ava-pb reads.fa reads.fa > overlaps.paf         # PacBio read overlap
./minimap2 -x ava-ont reads.fa reads.fa > overlaps.paf        # Nanopore read overlap
# man page for detailed command line options
man ./minimap2.1
```

# Mapping the reads – Minimap2

## Getting Started

```
git clone https://github.com/lh3/minimap2
cd minimap2 && make
# long sequences against a reference genome
./minimap2 -a test/MT-human.fa test/MT-orang.fa > test.sam
# create an index first and then map
./minimap2 -x map-ont -d MT-human-ont.mmi test/MT-human.fa
./minimap2 -a MT-human-ont.mmi test/MT-orang.fa > test.sam
# use presets (no test data)
./minimap2 -ax map-pb ref.fa pacbio.fq.gz > aln.sam          # PacBio CLR genomic reads
./minimap2 -ax map-ont ref.fa ont.fq.gz > aln.sam           # Oxford Nanopore genomic reads
./minimap2 -ax map-hifi ref.fa pacbio-ccs.fq.gz > aln.sam    # PacBio HiFi/CCS genomic reads (v2.1)
./minimap2 -ax lr:hq ref.fa ont-Q20.fq.gz > aln.sam         # Nanopore Q20 genomic reads (v2.27+)
./minimap2 -ax sr ref.fa read1.fa read2.fa > aln.sam        # short genomic paired-end reads
./minimap2 -ax splice ref.fa rna-reads.fa > aln.sam         # spliced long reads (strand unknown)
./minimap2 -ax splice -ur -k14 ref.fa reads.fa > aln.sam    # noisy Nanopore direct RNA-seq
./minimap2 -ax splice:hq -uf ref.fa query.fa > aln.sam      # PacBio Kinnex/Iso-seq (RNA-seq)
./minimap2 -ax splice --junc-bed=anno.bed12 ref.fa query.fa > aln.sam # use annotated junction
./minimap2 -ax splice:sr ref.fa r1.fq r2.fq > aln.sam       # short-read RNA-seq (v2.29+)
./minimap2 -ax splice:sr -j anno.bed12 ref.fa r1.fq r2.fq > aln.sam
./minimap2 -cx asm5 asm1.fa asm2.fa > aln.paf              # intra-species asm-to-asm alignment
./minimap2 -x ava-pb reads.fa reads.fa > overlaps.paf      # PacBio read overlap
./minimap2 -x ava-ont reads.fa reads.fa > overlaps.paf     # Nanopore read overlap
# man page for detailed command line options
man ./minimap2.1
```

# What's the script doing?

## Step 1: Indexing Reference Genome and Transcriptome

Create minimap2 indexes for efficient mapping.

*In Bash, including double quotes (") around variables is crucial for correct and safe handling of values, especially when they might contain spaces or special characters.*

```
minimap2 -d "$MINIMAP2_INDEX_TRANSCRIPTOME" "$TRANSCRIPTOME_FA"  
minimap2 -d "$MINIMAP2_INDEX_GENOME" "$GENOME_FA"
```

Each line is creating an index, one for the transcriptome and the other for the genome.

This speeds up future mapping, and it's key when using the same reference for multiple samples.



# What's the script doing?

*This FOR LOOP iterates over each of the `.fastq` files from the `$FASTQ_DIR`.*

```
for FASTQ_FILE in "$FASTQ_DIR"/*.fastq; do
  FILENAME=$(basename "$FASTQ_FILE" .fastq)

  # Transcriptome alignment
  minimap2 -t $THREADS -a -x map-ont "$MINIMAP2_INDEX_TRANSCRIPTOME" "$FASTQ_FILE" | \
  samtools view -Sb > "$BAM_DIR_TRANSCRIPTOME/$FILENAME.bam"

  # Genome alignment
  minimap2 -t $THREADS -a -x splice "$MINIMAP2_INDEX_GENOME" "$FASTQ_FILE" | \
  samtools view -Sb > "$BAM_DIR_GENOME/$FILENAME.bam"

  # Quantification with salmon
  salmon quant --ont -p $THREADS \
    -t "$TRANSCRIPTOME_FA" \
    -l U \
    -a "$BAM_DIR_TRANSCRIPTOME/$FILENAME.bam" \
    -o "$SALMON_DIR/$FILENAME"
done
```

# What's the script doing?

*This FOR LOOP iterates over each of the `.fastq` files from the `$FASTQ_DIR`.*

```
for FASTQ_FILE in "$FASTQ_DIR"/*.fastq; do  
    FILENAME=$(basename "$FASTQ_FILE" .fastq)
```

Iterates over each file on the directory ending in `.fastq`` and assigns the file name to the variable ``FASTQ_FILE``.

*The `"*"` is a wildcard, meaning "everything".*

Then, the extension is removed, and the file name is saved to the variable ``FILENAME``.

# What's the script doing?

```
# Transcriptome alignment
minimap2 -t $THREADS -a -x map-ont "$MINIMAP2_INDEX_TRANSCRIPTOME" "$FASTQ_FILE" | \
samtools view -Sb > "$BAM_DIR_TRANSCRIPTOME/$FILENAME.bam"
```

```
# Genome alignment
minimap2 -t $THREADS -a -x splice "$MINIMAP2_INDEX_GENOME" "$FASTQ_FILE" | \
samtools view -Sb > "$BAM_DIR_GENOME/$FILENAME.bam"
```

Now we are mapping (with minimap2) to the **reference transcriptome** and to the **reference genome**.

After each mapping, we use `samtools` to *compress* the **.SAM (Sequence Alignment/Map)** to a **.BAM (binary version of the Sequence Alignment/Map)**.

# What's the script doing?

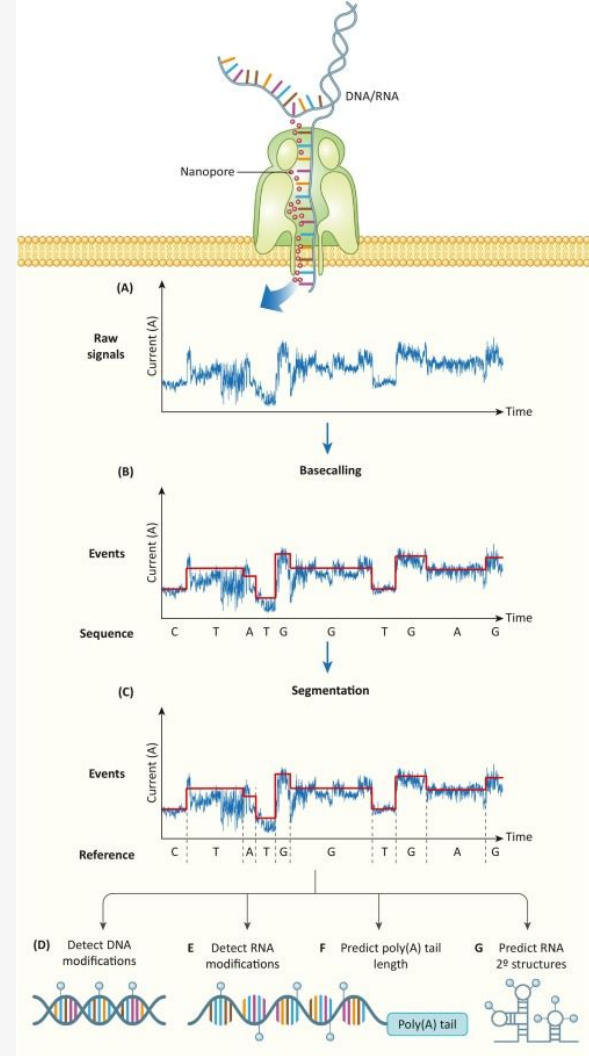
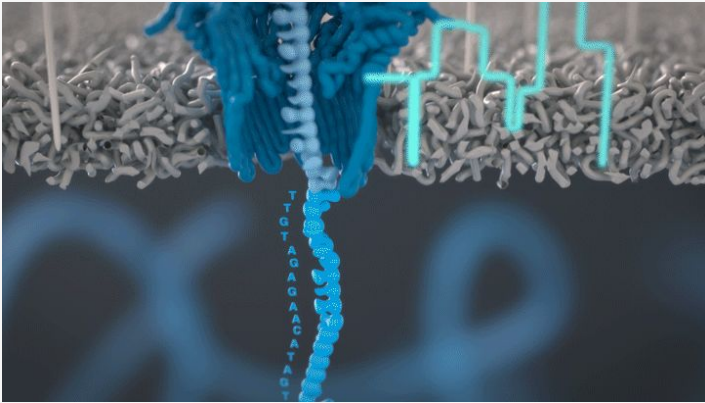
When using SALMON to quantify isoforms from long reads, a previous mapping is required (not necessary with illumina reads).

Here we quantify with salmon setting the parameters for `--ont` reads.

*`-l U` refers to unstranded library*

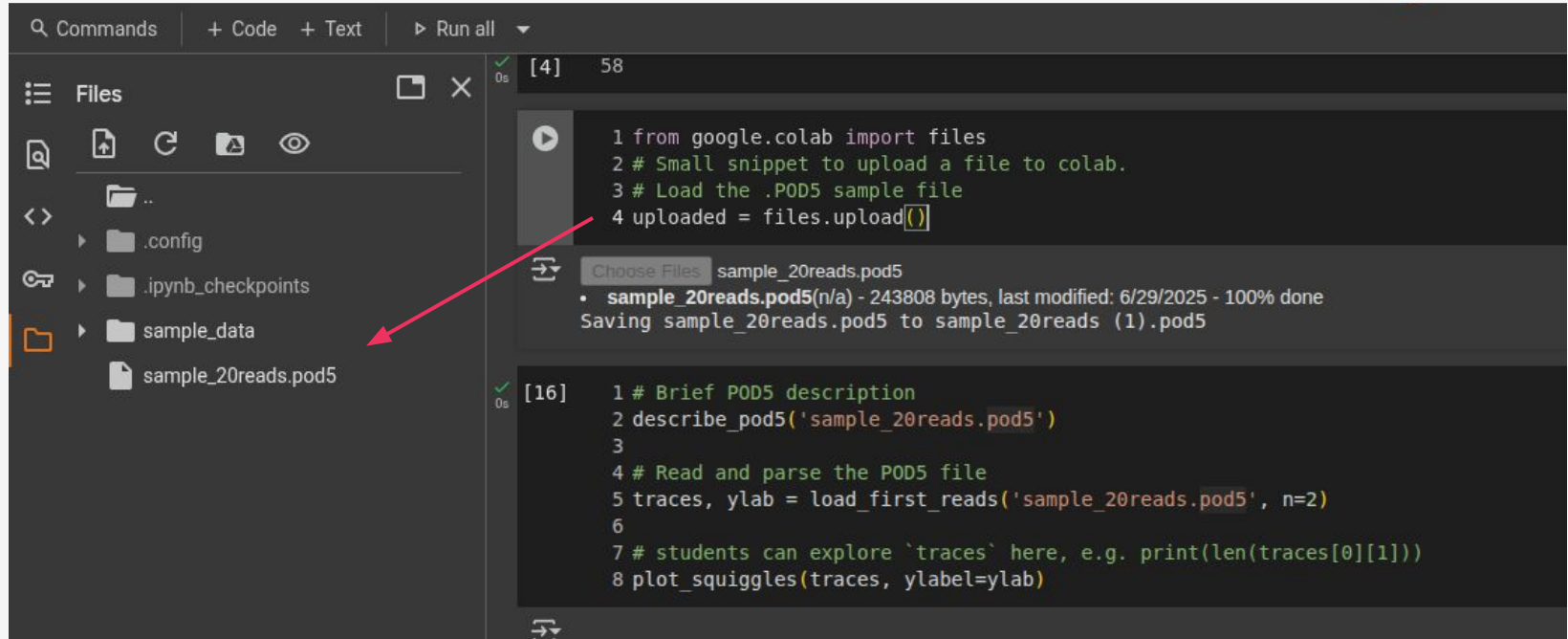
```
# Quantification with salmon
salmon quant --ont -p $THREADS \
  -t "$TRANSCRIPTOME_FA" \
  -l U \
  -a "$BAM_DIR_TRANSCRIPTOME/$FILENAME.bam" \
  -o "$SALMON_DIR/$FILENAME"
done
```

# How does Nanopores work?



# Comments on the code

Uploading files to work with Colab:



The screenshot displays the Google Colab interface. On the left, the 'Files' sidebar shows a directory structure with folders like '.config' and '.ipynb\_checkpoints', and files like 'sample\_data' and 'sample\_20reads.pod5'. A red arrow points from the 'sample\_20reads.pod5' file in the sidebar to the 'Choose Files' button in the code output area. The main code editor shows two code blocks. The first block, labeled '[4]', contains code to import 'files' from 'google.colab' and upload a file. The second block, labeled '[16]', contains code to describe and load the uploaded file. The output of the first block shows the file 'sample\_20reads.pod5' being saved as 'sample\_20reads (1).pod5'.

```
[4] 58
1 from google.colab import files
2 # Small snippet to upload a file to colab.
3 # Load the .POD5 sample file
4 uploaded = files.upload()
```

Choose Files sample\_20reads.pod5

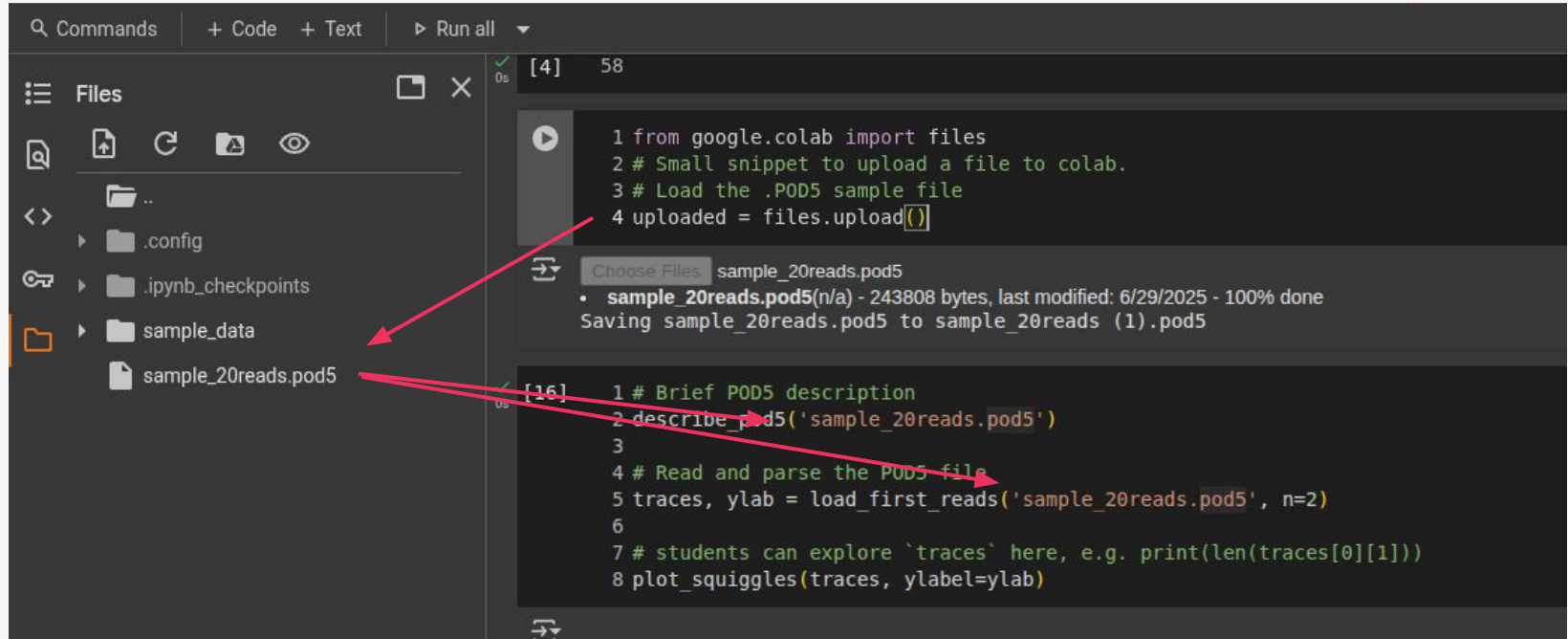
- sample\_20reads.pod5(n/a) - 243808 bytes, last modified: 6/29/2025 - 100% done

Saving sample\_20reads.pod5 to sample\_20reads (1).pod5

```
[16]
1 # Brief POD5 description
2 describe_pod5('sample_20reads.pod5')
3
4 # Read and parse the POD5 file
5 traces, ylab = load_first_reads('sample_20reads.pod5', n=2)
6
7 # students can explore `traces` here, e.g. print(len(traces[0][1]))
8 plot_squiggles(traces, ylabel=ylab)
```

# Comments on the code

Uploading files to work with Google Colab:



The screenshot shows the Google Colab interface. On the left, the 'Files' sidebar displays a directory structure with folders like '.config', '.ipynb\_checkpoints', and 'sample\_data', and a file named 'sample\_20reads.pod5'. A red arrow points from this file to the code cell below. The main area shows two code cells. The first cell, labeled '[4] 58', contains code to import 'files' from 'google.colab' and upload a file. Below the code is a 'Choose Files' dialog showing 'sample\_20reads.pod5' with its size (243808 bytes) and upload status (100% done). The second cell, labeled '[16] 0s', contains code to describe and load the uploaded file. A red arrow points from the 'sample\_20reads.pod5' file in the sidebar to the argument in the 'describe\_pod5' function call in the second code cell.

```
[4] 58
1 from google.colab import files
2 # Small snippet to upload a file to colab.
3 # Load the .POD5 sample file
4 uploaded = files.upload()
```

Choose Files sample\_20reads.pod5

- sample\_20reads.pod5(n/a) - 243808 bytes, last modified: 6/29/2025 - 100% done

Saving sample\_20reads.pod5 to sample\_20reads (1).pod5

```
[16] 0s
1 # Brief POD5 description
2 describe_pod5('sample_20reads.pod5')
3
4 # Read and parse the POD5 file
5 traces, ylab = load_first_reads('sample_20reads.pod5', n=2)
6
7 # students can explore `traces` here, e.g. print(len(traces[0][1]))
8 plot_squiggles(traces, ylabel=ylab)
```

# SALMON data

- **Salmon** is a **fast, alignment-free** (or quasi-alignment) tool for **quantifying transcript abundances** from RNA-Seq data.
- Uses **k-mers and lightweight mapping** to estimate expression without full alignment.
- **Bias-aware**: models fragment GC bias, positional bias, and sequence-specific bias to improve accuracy.
- Outputs **Transcripts Per Million (TPM)** per **isoform**.



# SALMON data

## How TPM is calculated:

Let:

$r_i$  = number of reads mapping to transcript  $i$

$l_i$  = length of transcript  $i$

1. Calculate Reads per Kilobase (RPK):

$$RPK_i = \frac{r_i}{l_i/1000}$$

2. Compute scaling factor (sum of RPKs):

$$\text{Scaling Factor} = \sum_j RPK_j$$

3. Compute TPM:

$$TPM_i = \frac{RPK_i}{\text{Scaling Factor}} \times 10^6$$



# SALMON data

## TPM – step 1: normalize for gene length

Original data:

Gene Name	Rep1 Counts	Rep2 Counts	Rep3 Counts
A (2kb)	10	12	30
B (4kb)	20	25	60
C (1kb)	5	8	15
D (10kb)	0	0	1

RPK – scaled by  
gene length:

Gene Name	Rep1 RPK	Rep2 RPK	Rep3 RPK
A (2kb)	5	6	15
B (4kb)	5	6.25	15
C (1kb)	5	8	15
D (10kb)	0	0	0.1



# SALMON data

TPM – step 2: normalize for sequencing depth

Gene Name	Rep1 RPK	Rep2 RPK	Rep3 RPK
A (2kb)	5	6	15
B (4kb)	5	6.25	15
C (1kb)	5	8	15
D (10kb)	0	0	0.1

Total RPK: 15 20.25 45.1

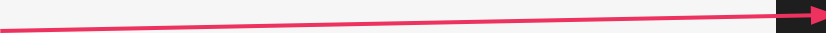
Tens of RPK: 1.5 2.025 4.51

TPM – scaled by  
gene length and  
sequencing  
depth (M):

Gene Name	Rep1 TPM	Rep2 TPM	Rep3 TPM
A (2kb)	3.33	2.96	3.326
B (4kb)	3.33	3.09	3.326
C (1kb)	3.33	3.95	3.326
D (10kb)	0	0	0.02

# SALMON OUTPUT

quant.sf



- ▶ colab\_quant\_files
- ▼ SQK-PCB109\_barcode01\_chr3
  - ▶ aux\_info
  - cmd\_info.json
  - ▶ libParams
  - ▶ logs
  - quant.sf
- ▶ SQK-PCB109\_barcode02\_chr3
- ▶ SQK-PCB109\_barcode03\_chr3
- ▶ SQK-PCB109\_barcode04\_chr3
- ▶ SQK-PCB109\_barcode05\_chr3

# SALMON OUTPUT

quant.sf

Name	Length	EffectiveLength	TPM	NumReads
AT3G01040.1	3098	100.000	0.000000	0.000
AT3G01040_P2	2969	100.000	0.000000	0.000
AT3G01040_P3	3095	100.000	166.362735	26.000
AT3G01050_JC6	2166	100.000	10.018608	1.566
AT3G01050_JS4	2373	100.000	11.724259	1.832
AT3G01050_c4	2306	100.000	125.424167	19.602
AT3G01060.2	1671	100.000	36.284498	5.671
AT3G01060.3	1670	100.000	0.000000	0.000
AT3G01060_P1	1695	100.000	123.679670	19.329
AT3G01070_P1	1116	100.000	19.195700	3.000
AT3G01080_P2	1801	100.000	0.000000	0.000
AT3G01090.1	2049	100.000	143.421496	22.415
AT3G01090_ID20	1632	100.000	0.000000	0.000
AT3G01090_ID23	2152	100.000	42.244833	6.602
AT3G01090_s1	2255	100.000	93.806271	14.661
AT3G01100_ID2	2841	100.000	0.000000	0.000
AT3G01100_JS4	3284	100.000	0.000000	0.000
AT3G01100_JS6	2928	100.000	131.786577	20.596
AT3G01100_P1	2753	100.000	0.000000	0.000
AT3G01100_P2	3049	100.000	17.444793	2.726
AT3G01100_P3	2829	100.000	0.000000	0.000
AT3G01120_P1	2461	100.000	1275.747279	199.380

# SALMON OUTPUT

## Name:

Isoform name from our transcriptome annotation.

## Length:

Length of the gene based on the reference.

## EffectiveLength:

Important in Illumina data – represents how many reads could fit along a transcript, adjusted for fragment size and biases.

## TPM:

Transcripts per million (normalized counts)

## NumReads:

How many reads Salmon estimates came from this transcript.

Name	Length	EffectiveLength	TPM	NumReads
AT3G01040.1	3098	100.000	0.000000	0.000
AT3G01040_P2	2969	100.000	0.000000	0.000
AT3G01040_P3	3095	100.000	166.362735	26.000
AT3G01050_JC6	2166	100.000	10.018608	1.566
AT3G01050_JS4	2373	100.000	11.724259	1.832
AT3G01050_c4	2306	100.000	125.424167	19.602
AT3G01060.2	1671	100.000	36.284498	5.671
AT3G01060.3	1670	100.000	0.000000	0.000
AT3G01060_P1	1695	100.000	123.679670	19.329
AT3G01070_P1	1116	100.000	19.195700	3.000
AT3G01080_P2	1801	100.000	0.000000	0.000
AT3G01090.1	2049	100.000	143.421496	22.415
AT3G01090_ID20	1632	100.000	0.000000	0.000
AT3G01090_ID23	2152	100.000	42.244833	6.602
AT3G01090_s1	2255	100.000	93.806271	14.661
AT3G01100_ID2	2841	100.000	0.000000	0.000
AT3G01100_JS4	3284	100.000	0.000000	0.000
AT3G01100_JS6	2928	100.000	131.786577	20.596
AT3G01100_P1	2753	100.000	0.000000	0.000
AT3G01100_P2	3049	100.000	17.444793	2.726
AT3G01100_P3	2829	100.000	0.000000	0.000
AT3G01120_P1	2461	100.000	1275.747279	199.380

# Comments on the code

```
1 import pandas as pd
2 from io import StringIO
3
4 # Build a new list for TPM-only dataframes
5 tpm_tables = []
6
7 for filename, filecontent in uploaded.items():
8     sample_name = filename.split("_")[1].replace(".sf", "") # Get the name from the filename
9     df = pd.read_csv(StringIO(filecontent.decode()), sep='\t') # Load each table as a pandas df
10    df = df.set_index('Name') # Set the isoform 'Name' as index
11
12    tpm = df[['TPM']].rename(columns={'TPM': sample_name}) # Keep only TPM column and rename it to sample name
13    tpm_tables.append(tpm)
14
15 tpm_combined = pd.concat(tpm_tables, axis=1) # Merge all dataframes (TPM)
16 tpm_combined["Gene"] = tpm_combined.index.str.slice(0, 9) # Extract gene name for later grouping
17
18 # Sorting the columns (just for clearer prints) :)
19 barcode_order = [
20     "barcode01", "barcode02", "barcode03", "barcode04", "barcode05", "barcode06",
21     "barcode07", "barcode08", "barcode09", "barcode10", "barcode11", "barcode12"
22 ]
23 sorted_columns = ['Gene'] + barcode_order
24 tpm_combined = tpm_combined[sorted_columns]
25
26 tpm_combined.head() # .head() command prints the first rows
```

# *Arabidopsis* transcriptome notation

AT3G61860.1

AT3G61860.1

Chromosome number – Gene number  
+ isoform suffix








# SUPPA2

Isoform proportions



# SUPPA2 Workflow – 3 Key Steps

1.  **Generate alternative splicing events** from a GTF annotation  
➤ `suppa.py generateEvents -i annotation.gtf -o events -f ioi`
2.  **Calculate PSI ( $\Psi$ ) values** from transcript-level expression (e.g., TPMs)  
➤ `suppa.py psiPerIsoform -i transcripts.ioi -e sample1.tpm,sample2.tpm -o psi_values.psi`
3.  **Compare conditions to get  $\Delta$ PSI and p-values**  
➤ `suppa.py diffSplice -m empirical -gc  
-i transcripts.ioi  
--psi condition1.psi condition2.psi  
--tpm condition1.tpm condition2.tpm  
-o diff_splicing`

# SUPPA 2 – In a nutshell

PSI ( $\Psi$ ) and delta PSI ( $\Delta\Psi$ )

GENE	ISOFORM	TPM (cond 1)	PSI (cond 1)	TPM (cond 2)	PSI (cond 2)	Delta PSI
Gene_A	Isoform_1	10	0.1	50	0.625	+ 0.525
Gene_A	Isoform_2	70	0.7	30	0.375	- 0.325
Gene_A	Isoform_3	20	0.2	0	0	- 0.2

# Setting some filters

## ✓ Creating a filter for significant values

We will set 2 thresholds:

- `dpsi_cut`: We will consider that variation on the psi less than 0.1 are irrelevant (or noise)
- `p-value_cut`: Only keep the changes that are significant ( $< 0.05$ )

```
1 # --- All the comparisons made -----
2 contrasts = [
3     "Nuc_light-Nuc_dark",
4     "Nuc_light-Cit_light",
5     "Nuc_light-Cit_dark",
6     "Nuc_dark-Cit_light",
7     "Nuc_dark-Cit_dark",
8     "Cit_light-Cit_dark",
9 ]
10 dpsi_cut = 0.10      # |ΔPSI| threshold
11 p_cut    = 0.05      # raw p threshold (see §8 for FDR)
12
13 # -----
14
15 sig = {
16     c: (
17         df_suppa[f"{c}_p-val"] <= p_cut
18     ) & (
19         df_suppa[f"{c}_dPSI"].abs() >= dpsi_cut
20     ) & (
21         df_suppa[f"{c}_dPSI"].abs() < 1      # We also want to remove the
22     )
23     for c in contrasts
24 }
25
26 '''
27 This filter can be improved by filtering the GENES with at least one
28 significant value change. (you can try it!)
29 Suggestion: use .groupby() and take advantage of the multiindex
30 '''
```

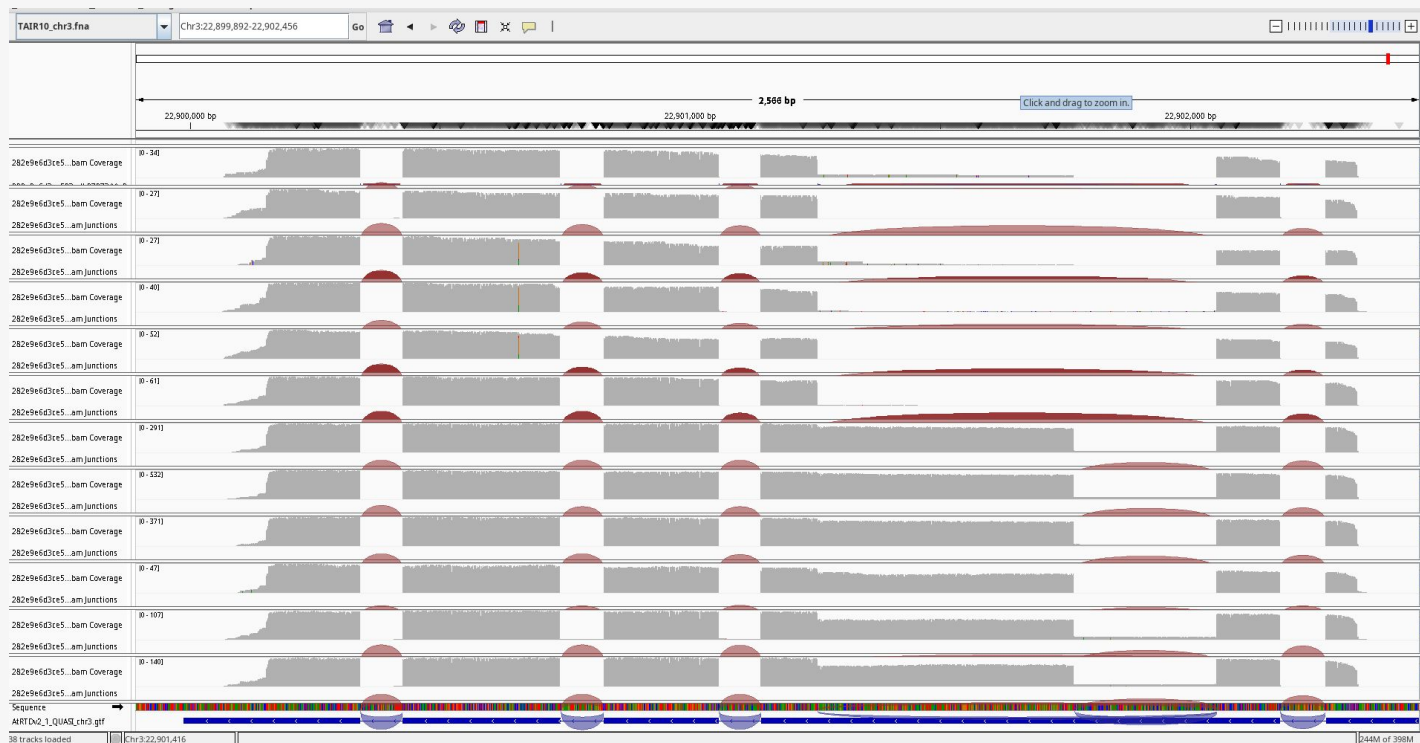
# Visualizing isoforms

We can see the isoforms here: <https://boxify.boku.ac.at/>



<https://boxify.boku.ac.at/>

# Integrative Genomics Viewer – IGV



<https://igv.org/>